Mimetic finite difference methods for diffusion equations on AMR meshes

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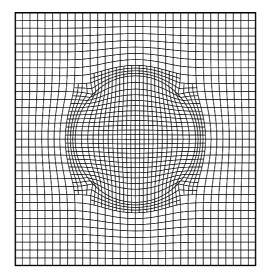
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- Recent developments nad future plans
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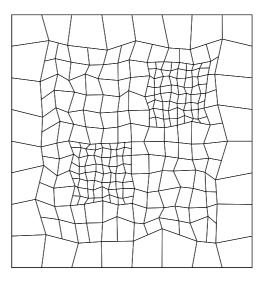


Objectives

What are the perfect discretizations?

- they preserve and mimic mathematical properties of physical systems;
- they are accurate on adaptive smooth and non-smooth grids;



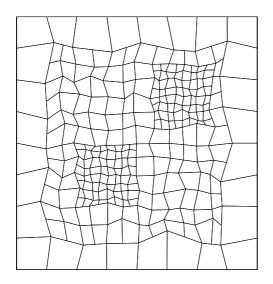


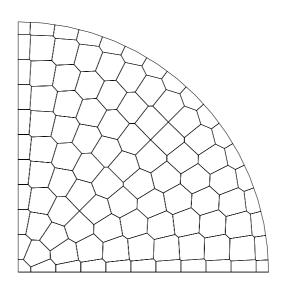


Objectives

What are the perfect discretizations?

- they preserve and mimic mathematical properties of physical systems;
- they are accurate on adaptive smooth and non-smooth grids;
- they can be used for a wide family of grids and operators.







Model diffusion problem

We consider the elliptic equation

$$-\text{div}(\boldsymbol{K}\,\nabla p) = b \qquad \text{in} \qquad \Omega$$

subject to the homogeneous Dirichlet b.c.

$$p=0$$
 on $\partial\Omega$.

The problem can be reformulated as a system of first order equations:

$$div \mathbf{f} = b,$$

$$\mathbf{f} = -\mathbf{K} \nabla p.$$

For simplicity we assume that K = I.



Support operator method (1/2)

Consider the mathematical identity:

$$\int_{\Omega} \operatorname{grad} p \, \boldsymbol{f} \, \mathrm{d}x = -\int_{\Omega} \operatorname{div} \boldsymbol{f} \, p \, \mathrm{d}x \qquad \forall \boldsymbol{f} \in H_{div}(\Omega), \ p \in H_0^1(\Omega).$$

Support-operators (SO) methodology (for div & grad):

- 1. define degrees of freedom for variables p and f;
- 2. equip the discrete spaces for p and f with scalar products $[\cdot, \cdot]_Q$ and $[\cdot, \cdot]_X$, respectively;
- 3. choose a discrete approximation to the divergence operator, the *prime* operator $DIV: X_d \rightarrow Q_d$;
- 4. derive the discrete approximation of the gradient operator, the *derived* operator GRAD: $Q_d \rightarrow X_d$, from the discrete Green formula:

$$[f^d, \operatorname{GRAD} p^d]_X = -[\operatorname{DIV} f^d, p^d]_Q \qquad \forall p^d \in Q_d, f^d \in X_d.$$



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Support operator method (2/2)

Applications of the SO methodology include:

■ Electromagnetics: discrete operators DIV, GRAD, CURL and CURL mimic:

$$\operatorname{div} \operatorname{curl} = 0, \quad \operatorname{curl} \operatorname{grad} = 0$$
$$\int_{\Omega} \operatorname{curl} \boldsymbol{E} \cdot \boldsymbol{H} \, \mathrm{d}x = \int_{\Omega} \operatorname{curl} \boldsymbol{H} \cdot \boldsymbol{E} \, \mathrm{d}x + \oint_{\partial \Omega} (\boldsymbol{E} \times \boldsymbol{H}) \cdot \boldsymbol{n} \, \mathrm{d}s$$

■ CFD: discrete operators DIV and GRAD mimic:

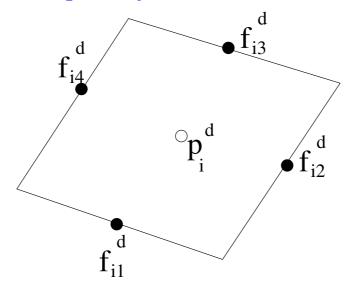
$$\int_{\Omega} \operatorname{grad} \boldsymbol{u} \colon \boldsymbol{T} \, \mathrm{d}x = -\int_{\Omega} \operatorname{div} \boldsymbol{T} \cdot \boldsymbol{u} \, \mathrm{d}x + \oint_{\partial \Omega} \boldsymbol{u} \cdot (\boldsymbol{T} \cdot \boldsymbol{n}) \, \mathrm{d}s$$

■ Gas dynamics, poroelasticity, magnetic diffusion, radiation diffusion, etc...

http://www.sci.sdsu.edu/compscims/MIMETIC/index.htm



Step 1 (degrees of freedom for p and f).



- lacksquare p_i^d is defined at a center of cell e_i .
- $f_{i1}^d, \ldots, f_{i4}^d$ are defined at mid-points of cell edges. They approximate the normal components of f, e.g.

$$f_{i1}^d pprox m{f} \cdot m{n}_{i1}.$$

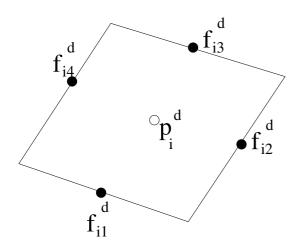


Step 2 (scalar products for p^d and f^d).

Let Q_d be a vector space of discrete intencities with the scalar product

$$[p^d, q^d]_Q = \sum_{i=1}^N |e_i| p_i^d q_i^d \approx \int_{\Omega} p(x) q(x) dx.$$

Let X_d be a vector space of discrete fluxes with a scalar product $[f^d, g^d]_X \approx \int_{\Omega} \boldsymbol{f}(x) \cdot \boldsymbol{g}(x) \, \mathrm{d}x.$



$$[f_{i2}^d \qquad [f_i^d, g_i^d]_{X_{e_i}} = rac{1}{2} \sum_{j=1}^4 |T_{ij}| \, m{f}_{ij}^d \cdot m{g}_{ij}^d \, .$$

Then
$$[f^d, g^d]_X = \sum_{i=1}^N [f^d_i, g^d_i]_{X_{e_i}}$$
.

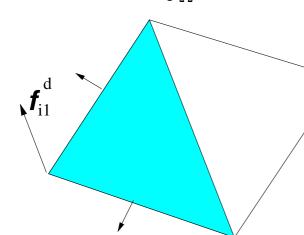


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$$[f_i^d, g_i^d]_{X_{e_i}} = \frac{1}{2} \sum_{j=1}^4 |T_{ij}| \, \boldsymbol{f}_{ij}^d \cdot \boldsymbol{g}_{ij}^d$$

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$$[f^d, g^d]_X = \sum_{i=1}^N [f_i^d, g_i^d]_{X_{e_i}}$$
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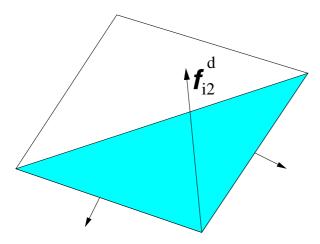


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Then
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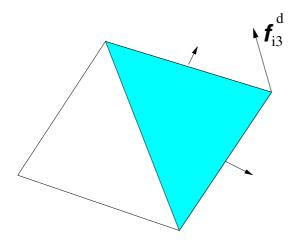


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$$[f_i^d, g_i^d]_{X_{e_i}} = \frac{1}{2} \sum_{j=1}^4 |T_{ij}| \, \boldsymbol{f}_{ij}^d \cdot \boldsymbol{g}_{ij}^d$$

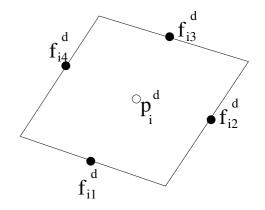
Then
$$[f^d, g^d]_X = \sum_{i=1}^N [f_i^d, g_i^d]_{X_{e_i}}$$
.



Steps 3 & 4 (prime and derived operators).

The prime operator DIV follows from the Gauss theorem:

$$\operatorname{div} \boldsymbol{f} = \lim_{|e| \to 0} \frac{1}{|e|} \oint_{\partial e} \boldsymbol{f} \cdot \boldsymbol{n} \ dl.$$



Center-point quadrature gives

$$\left(\mathbf{DIV}f^{d}\right)_{i} = \frac{1}{|e_{i}|} \left(f_{i1}^{d} |l_{1}| + f_{i2}^{d} |l_{2}| + f_{i3}^{d} |l_{3}| + f_{i4}^{d} |l_{4}| \right)$$

The derived operator GRAD is implicitly given by

$$[f^d, \operatorname{GRAD} p^d]_X = -[\operatorname{DIV} f^d, p^d]_Q \qquad \forall p^d \in Q_d, f^d \in X_d.$$



Short summary.

The stationary diffusion problem

$$-\operatorname{div} \boldsymbol{K} \nabla p = b \quad \text{in } \Omega$$
$$p = 0 \quad \text{on } \partial \Omega$$

is rewritten as the 1st order system

$$f = -K\nabla p$$
, $\operatorname{div} f = b$

and discretized as follows:

$$f^d = -\text{GRAD}\,p^d, \qquad \text{DIV}\,f^d = b^d.$$



By the definition,

$$[f^d, \operatorname{GRAD} p^d]_X = -[\operatorname{DIV} f^d, p^d]_Q.$$

Let $\langle \cdot, \cdot \rangle$ be the usual vector dot product. Then

$$[p^d, q^d]_Q = \langle \mathcal{D}p^d, q^d \rangle, \qquad [f^d, g^d]_X = \langle \mathcal{M}f^d, g^d \rangle.$$

Combining the last two formulas, we get

$$\begin{split} [f^d, \operatorname{GRAD} p^d]_X &= &< \mathcal{M} \, f^d, \operatorname{GRAD} p^d > \\ &= &- [\operatorname{DIV} f^d, \, p^d]_Q = - < f^d, \operatorname{DIV}^t \mathcal{D} \, p^d >. \end{split}$$

Therefore,

$$GRAD = -\mathcal{M}^{-1}DIV^{t}\mathcal{D}.$$



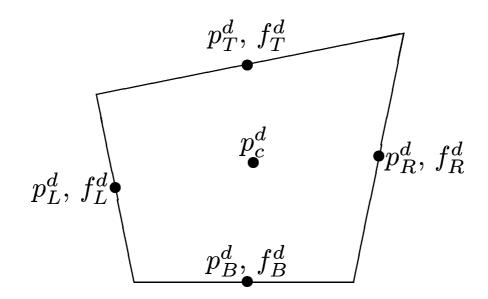
A local SO method mimics the mathematical identity

$$\int_{e} \mathbf{f} \cdot \operatorname{grad} p \, dx + \int_{e} \operatorname{div} \mathbf{f} \, p \, dx = \int_{\partial e} p \, \mathbf{f} \cdot \mathbf{n} \, dl.$$

Step 1 (degrees of freedom):

 p^d : at cell centers and edge centers

 f^d : normal components at edge centers





Steps 2 & 3 (discrete identity and prime operator).

The prime operator DIV is derived from the Gauss theorem:

$$DIV f^d = \frac{1}{|e|} \left(f_R^d |l_R| + f_T^d |l_T| + f_L^d |l_L| + f_B^d |l_B| \right)$$

Derivation of the discrete identity:

$$\int_{e} \mathbf{f} \cdot \operatorname{grad} p \, \mathrm{d} x \approx [f^d, \operatorname{GRAD} p^d]_{X_e}$$

$$\int \operatorname{div} \boldsymbol{f} \, p \, \mathrm{d}x \approx (\operatorname{DIV} f^d) \, p_c^d \, |e|$$

$$\int_{\partial e} p \, \boldsymbol{f} \cdot \boldsymbol{n} \, \mathrm{d}l \approx p_R^d f_R^d |l_R| + p_T^d f_T^d |l_T| + p_L^d f_L^d |l_L| + p_B^d f_B^d |l_B|$$



Step 4 (derived operator).

Replacing integrals in the Green formula by their approximations, we get

$$\operatorname{GRAD} p^d = \mathcal{M}_e^{-1} \left(egin{array}{c} |l_R|(p_R^d - p_c^d) \ |l_T|(p_T^d - p_c^d) \ |l_L|(p_L^d - p_c^d) \ |l_B|(p_B^d - p_c^d) \end{array}
ight)$$

where

$$<\mathcal{M}_e f^d, g^d> = [f^d, g^d]_{X_e}$$

and $f^d = (f_R^d, f_T^d, f_L^d, f_B^d)^t$. The local discretization reads

$$DIV f^d = b^d,$$

$$f^d = -GRAD p^d.$$

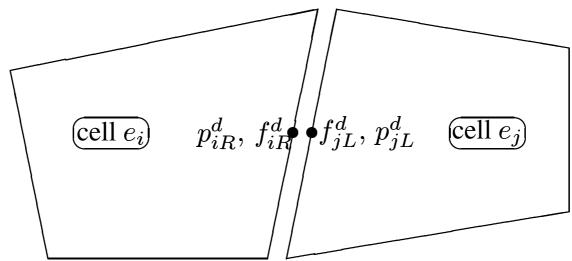


Short summary.

- $\blacksquare \text{ matrix } M_e^{-1} \in \Re^{4 \times 4};$
- discrete divergence and gradient operators mimic essential properties of the continuous operators (local mass conservation, Green formula);
- discretization and continuity conditions are separated;
- we do not assume anything about a grid structure.



$$\int_{\partial e} p \, \boldsymbol{f} \cdot \boldsymbol{n} \, \mathrm{d}l \approx p_R^d f_R^d |l_R| + p_T^d f_T^d |l_T| + p_L^d f_L^d |l_L| + p_B^d f_B^d |l_B|.$$



The global discretization is achieved by imposing the continuity of fluxes

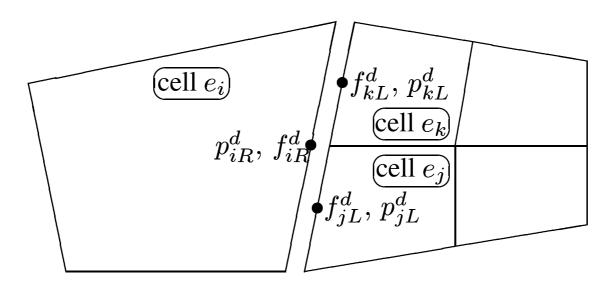
$$f_{iR}^d = -f_{jL}^d$$

and interface intensities

$$p_{iR}^d = p_{jL}^d.$$



Locally refined meshes (1/6)



The global discretization is achieved by imposing the continuity of fluxes

$$f_{iR}^d = -f_{jL}^d = -f_{kL}^d$$

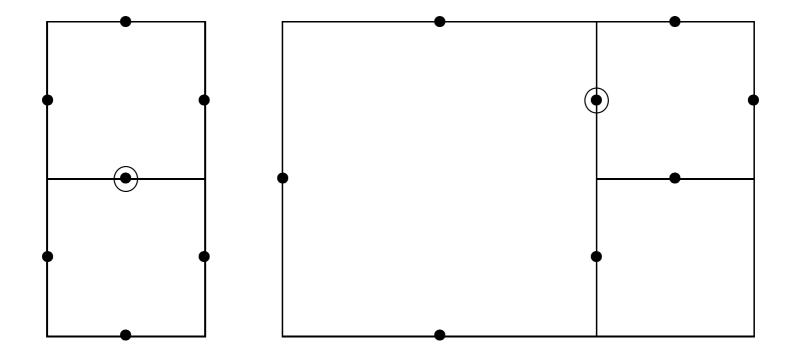
and interface intensities

$$|l_{iR}| p_{iR}^d = |l_{jL}| p_{jL}^d + |l_{kL}| p_{kL}^d.$$



Locally refined meshes (2/6)

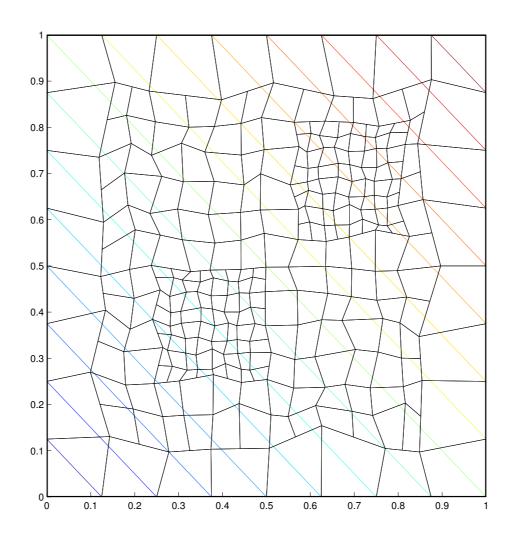
Stencils of a stiffness matrix for interface intencities.





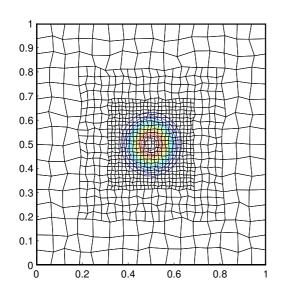
Locally refined meshes (3/6)

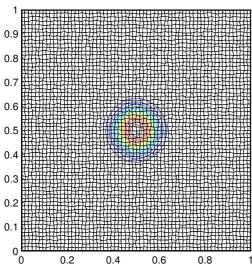
The derived mimetic discretizations are exact for linear solutions.





Locally refined meshes (4/6)

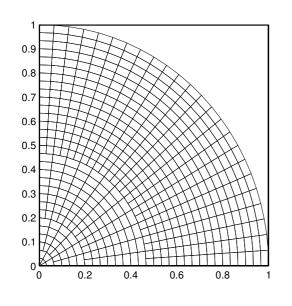


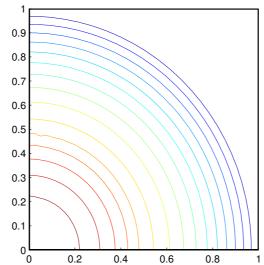


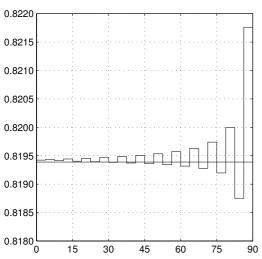
l	N	$arepsilon_p$	$arepsilon oldsymbol{f}$	#itr	CPU,s	
	AMR grids					
0	256	7.00e-2	8.18e-2	12	0.05	
1	556	1.64e-2	3.42e-2	15	0.14	
2	988	3.74e-3	1.74e-2	16	0.28	
3	3952	9.96e-4	7.57e-3	16	1.33	
4	<u>15808</u>	2.40e-4	3.79e-3	17	6.21	
	Uniform grids					
0	256	7.00e-2	8.18e-2	12	0.05	
1	1024	1.79e-2	3.40e-2	13	0.27	
2	4096	3.91e-3	1.62e-2	14	1.25	
3	16384	9.44e-4	7.30e-3	15	5.58	
4	<u>65536</u>	2.32e-4	3.76e-3	17	25.3	
$p(x, y) = 1 - \tanh\left(\frac{(x - 0.5)^2 + (y - 0.5)^2}{0.01}\right).$						

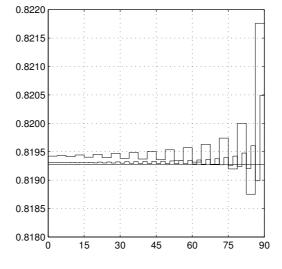


Locally refined meshes (5/6)









Spherically symmetric problem in r-z coordinates with the exact solution:

$$p(R) = \frac{553}{640} - \frac{R^2}{6} - \frac{R^4}{20}$$

when R < 0.5 and

$$p(R) = \frac{101}{120} - \frac{R^2}{12} - \frac{R^4}{40}$$

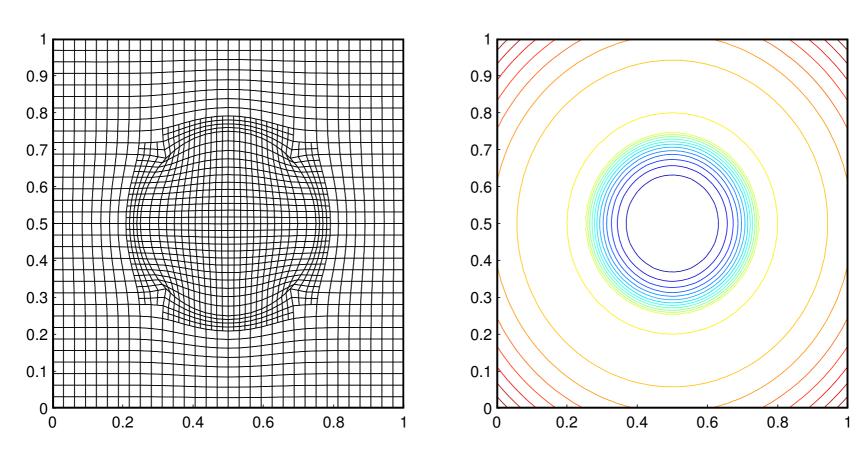
when 0.5 < R < 1.



Locally refined meshes (6/6)

Let us consider the diffusion problem with strong material discontinuity

$$[K] = 100$$
 at $\sqrt{(x-0.5)^2 + (y-0.5)^2} = 0.25$.



The locally fitted grids were generated by Shengtai Li (T-7).



SO and mixed FE methods (1/3)

The system of mimetic finite difference equations

$$f^d = -\text{GRAD}\,p^d, \qquad \text{DIV}\,f^d = b^d$$

can be rewritten as

$$[f^d, g^d]_X + [GRADp^d, g^d]_X = 0,$$

 $[DIV f^d, q^d]_Q = [b^d, q^d]_Q.$

Recall that by the definition,

$$[f^d, \, \operatorname{GRAD} p^d]_X = -[\operatorname{DIV} f^d, \, p^d]_Q.$$



SO and mixed FE methods (2/3)

Thus, the mimetic discretizations are equivalent to

$$[f^d, g^d]_X - [DIV f^d, p^d]_Q = 0,$$

 $-[DIV f^d, q^d]_Q = -[b^d, q^d]_Q, \quad \forall p^d \in Q_d, g^d \in X_d.$

On the other hand, the MFE method with the *Raviart-Thomas* elements gives

$$(f^h, \mathbf{g}^h) - (\operatorname{div} f^h, p^h) = 0,$$

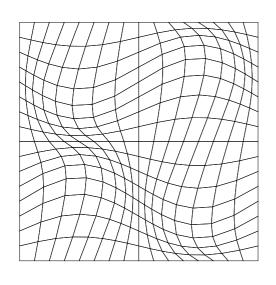
$$-(\operatorname{div} f^h, q^h) = -(b, q^h) \quad \forall q^h \in Q_h, g^h \in X_h.$$

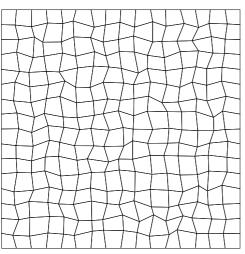
 p^d : at cell centers one per cell

Degrees of freedom: f^d : normal components normal components, at edge centers one per edge



SO and mixed FE methods (3/3)





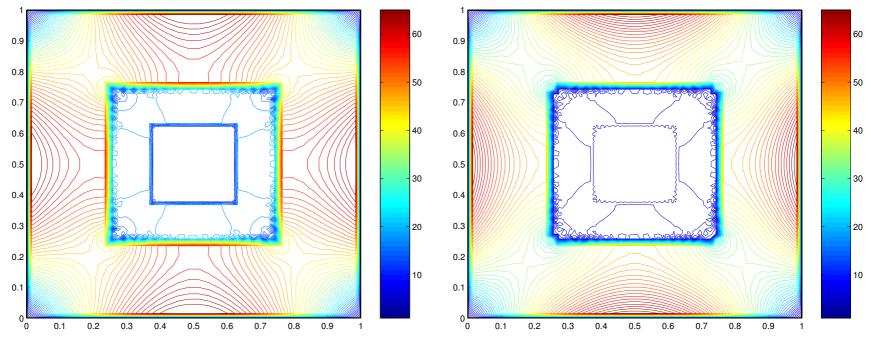
h^{-1}	modified	d RT FE	SO FD	
	$arepsilon_p$	$arepsilon_{m{f}}$	$arepsilon_p$	$arepsilon oldsymbol{arepsilon}$
16	1.58e-3	2.34e-2	1.61e-3	2.35e-2
32	7.95e-4	1.22e-2	7.99e-4	1.22e-2
64	3.98e-4	6.29e-3	3.99e-4	6.29e-3
128	1.99e-4	3.22e-3	1.99e-4	3.22e-3
256	9.97e-5	1.64e-3	9.97e-5	1.64e-3
512	4.98e-5	8.32e-4	4.98e-5	8.32e-4
	$arepsilon_p$	$arepsilon_{m{f}}$	$arepsilon_p$	$arepsilon_{oldsymbol{f}}$
16	1.42e-3	2.24e-2	1.43e-3	2.25e-2
32	7.15e-4	1.17e-2	7.18e-4	1.17e-2
64	3.59e-4	5.96e-3	3.59e-4	5.98e-3
128	1.80e-4	3.06e-3	1.80e-4	3.07e-3
256	9.00e-5	1.56e-3	9.00e-5	1.56e-3
512	4.50e-5	7.93e-4	4.50e-5	7.93e-4



SO and FD methods (1/1)

In collaboration with M.Pernice (CCS-3), the SO method was compared with the FD method by R.Ewing, R.Lazarov, and P.Vassilevki (1991):

- the FD method works on rectangular locally refined grids;
- in the case of smooth solutions, the FD method results in larger error (left picture) on irregular grid interfaces:





SO and CV methods (1/1)

The control-volume mixed FE method by T.Russell (2001):

- the method does not preserve the uniform flow on irregular grids;
- the principle difficulty is the scalar product in a space of fluxes.

The control-volume method on general polygonal meshes by T.Palmer (2001):

- the method is exact for linear solutions;
- the method results in non-symmetric matrices.

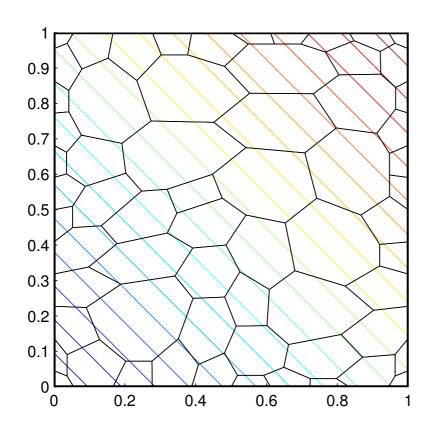
The SO method on general polygonal meshes (2003):

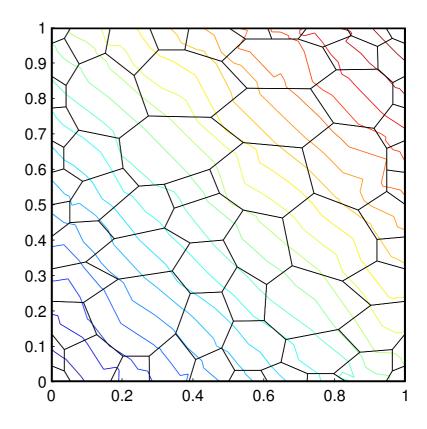
- the method is exact for linear solutions;
- the method results in symmetric positive definite matrices.



Recent developments (1/3)

Exact solution is p(x, y) = x + y. A new scalar product in the space of fluxes results in mimetic discretizations which are exact for linear solutions.





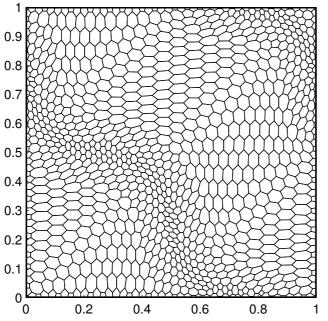
The polygonal grids were generated by Raphael Loubere (T-7).

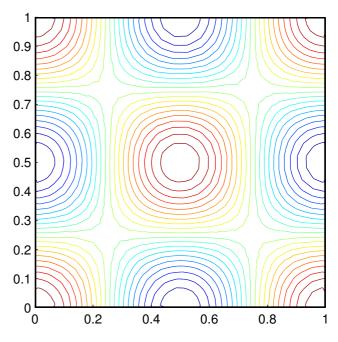


Recent developments (2/3)

Convergence test for exact solution $p(x, y) = \sin(2\pi x)\sin(2\pi y)$.

m	New scalar product		Old scalar product	
	$arepsilon_p$	$arepsilon_f$	$arepsilon_p$	$arepsilon_f$
166	1.07e-1	3.68e-1	1.81e-1	4.57e-1
598	2.60e-2	1.64e-1	3.39e-2	2.52e-1
2230	5.11e-3	8.28e-2	6.64e-3	1.72e-1
8566	1.05e-3	4.29e-2	1.51e-3	1.20e-1

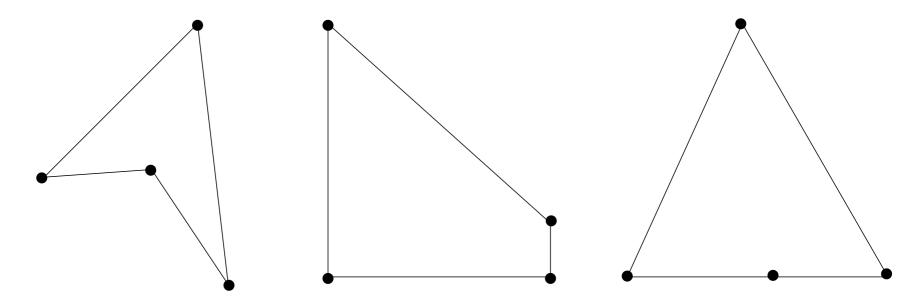






Recent developments (3/3)

Examples of bad-shaped elements which are common for locally refined and non-matching meshes:



We believe that the new methodology can be extended to all the above elements.



Conclusion

- the convergence of mimetic discretizations for the linear diffusion equation is optimal on locally refined meshes in both Cartesian and r-z gemetries (2nd order on smooth meshes but only 1st order for fluxes on random grids);
- the mimetic discretizations are comparable with mixed FE discretizations but more preferable than the discretizations based on CV or FD methods;
- a reduced system for interface intencities has SPD coefficient matrix and can be efficiently solved with a PCG method;
- the preliminary numerical experiments on general polygonal meshes show the optimal convergence rate for mimetic discretizations (2nd order for intencities and 1st order for fluxes).

